

Lecture 6: Sampling

ECE 401: Signal and Image Analysis

University of Illinois

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- 1 Spectrogram Review
- 2 The Sampling Problem: 2π Ambiguity
- 3 Fourier Series

Outline

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- 2 The Sampling Problem: 2π Ambiguity
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On-Board Practice

Eastern Illinois is famous for two types of frogs. The Monotonal Illinois Frog (MIF) has a sinusoidal call at 50Hz:

$$b(t) = \sin(100\pi t)$$

The Bitonal Illinois Frog (BIF) has a call which is two sinusoids, at 49Hz and 51Hz:

$$m(t) = 0.5 \sin(98\pi t) + 0.5 \sin(102\pi t)$$

You have a signal; in order to determine if it's $b(t)$ or $m(t)$, you compute its spectrogram. How long does the spectrogram window need to be?

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Definition of Sampling

“Sampling” means that you measure the signal (sound pressure, voltage, or whatever) once per $T = \frac{1}{F_s}$ seconds, and store the value in a computer.

$$x[n] = x(nT) = x(n/F_s)$$



2π Ambiguity in Phase

The problem with sampling is that $\cos(2\pi) = \cos(0)$. More generally,

$$e^{j\theta+2\pi k} = e^{j\theta} e^{j2\pi k} = e^{j\theta} \quad \text{for any integer } k$$

2π Ambiguity in Frequency

The 2π -ambiguity means that

$$\cos(\omega n) = \cos((\omega + 2\pi k)n) \quad \text{for any integer } k$$

- **Every sample** of the signal on the left is exactly the same as **every sample** of the signal on the right.
- The only way to tell them apart would be if you knew the function in the spaces between the samples, e.g., at $n = \frac{1}{2}$.
- But you don't. The integer values of n are all you get.
- So in this sense, the frequencies ω and $\omega \pm 2\pi$ are **exactly the same frequency**.

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Spectrum of a Discrete-Time Signal is Periodic

In order to represent the fact that ω and $\omega + 2\pi$ are actually the same angle, we usually say that **the spectrum of every discrete-time signal is “periodic” with period 2π :**

$$X(\omega) = X(\omega + 2\pi)$$

Spectrum of a Discrete-Time Signal is Periodic

For example,

$x[n] = \sin(0.137\pi n) = (\exp(0.137j\pi n) + \exp(-0.137j\pi n))/2j$ has the Fourier-series based power spectrum

$$|X_\omega|^2 = \begin{cases} \frac{1}{4} & \omega = 0.137\pi + 2\pi k \text{ for any integer } k \\ \frac{1}{4} & \omega = -0.137\pi + 2\pi k \text{ for any integer } k \\ 0 & \text{otherwise} \end{cases}$$

Often we will leave off the $2\pi k$, and assume that you understand it's still there. For example, the following equation has exactly the same meaning as the one above, because $X(\omega)$ is periodic:

$$|X_\omega|^2 = \begin{cases} \frac{1}{4} & \omega = 0.137\pi \\ \frac{1}{4} & \omega = -0.137\pi \\ 0 & \text{otherwise} \end{cases}$$

Principal Phase: Example

It usually makes most sense to define ω in the range $[-\pi, \pi]$.
Thus, for example, an 875Hz tone

$$x(t) = \sin(1750\pi t)$$

when sampled at $F_s = 1000$ samples/second produces

$$\begin{aligned}x[n] &= \sin\left(\frac{1750\pi n}{1000}\right) = \sin(1.75\pi n) \\ &= \sin(-0.25\pi n) = -\sin(0.25\pi n)\end{aligned}$$

Principal Phase in Principle

So if the continuous-time frequency of a sine wave is Ω radians/second, then its discrete-time frequency can be defined in a few different ways:

Ambiguous definition: $\omega = \frac{\Omega}{F_s} \pm 2\pi k$ for any integer k

Unambiguous definition: $\omega = \text{mod} \left(\frac{\Omega}{F_s}, 2\pi \right)$

where the modulo operator is usually defined so that $-\pi < \omega \leq \pi$, but sometimes we'll wrap it around and define it so that $0 \leq \omega < 2\pi$.

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Fourier Series

Remember that any periodic signal can be written as

$$x(t) = \sum_{k=-\infty}^{\infty} X_k e^{jk\Omega_0 t}$$

So when sampled, it would be

$$x[n] = \sum_{k=-\infty}^{\infty} X_k e^{jk\omega_0 n}, \quad \omega_0 = \frac{\Omega_0}{F_s}$$

Except that formula is **INCORRECT** in the sense that, for large values of k , $k\omega_0 > \pi$! (I hate to call it incorrect; DECEPTIVE would be a better term. The signal really is $e^{jk\omega_0 n}$, but the signal is ALSO $e^{j(k\omega_0 - 2\pi)n}$).

Aliasing

Problem: the modulo operator “wraps around” all of the high-frequency components, back into the principal phase range:

$$x[n] = \sum_{k=-\infty}^{\infty} X_k e^{j\text{mod}(k\omega_0, 2\pi)n}$$

Since the summation is infinite, that means we have an infinite number of high-frequency harmonics wrapped around into the principal phase area. This means the signal is distorted.

Solution: Lowpass Filtering

The solution is to lowpass filter $x(t)$ (i.e., smooth it) before sampling it.

We lowpass filter it in order to zero-out any harmonics with frequency $k\Omega_0 > \pi F_s$, equivalent to $kF_0 > 0.5F_s$. Thus

$$y(t) = \text{LPF}(x(t)) = \sum_{-k_{max}}^{k_{max}} X_k e^{jk\Omega_0 t}$$

where $k_{max} = F_s/2F_0$. Then

$$y[n] = y(n/F_s) = \sum_{-k_{max}}^{k_{max}} X_k e^{jk\omega_0 n}$$

where we now have the guarantee that

$$-\pi \leq -k_{max}\omega_0 \quad \text{and} \quad k_{max}\omega_0 \leq \pi$$

Nyquist Theorem

It is possible to reconstruct a signal from its samples as long as its highest frequency component has a frequency $F < 0.5F_s$.